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# APPLIED MATHEMATICS AND STATISTICS LABORATORY STANFORD UNIVERSITY CALIFORNIA

MINIMAX PROCEDURES FOR TWO-VALUED DECISION PROBLEMS WHEN THE SIZE OF SAMPLE IS FIXED

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S. G. ALLEN, JR.

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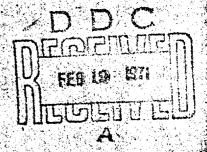
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PRELIMINARY DRAFT For Comment



MINIMAX PROCEDURES FOR TWO-VALUED DECISION PROBLEMS WHEN THE SIZE OF SAMPLE IS FIXED

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#### S. G. Allen, Jr.

1. The Minimax Solution for the Case of a Simple Dichotomy. The problem to be considered is a statistical decision procedure for choosing between two alternative actions,  $A_1$  and  $A_2$ , after taking n independent observations on a random variable  $\pi$ . The probability density p(x,0) of x is known except for the value of a real parameter  $\theta$  which is assumed to be one-dimensional. The parameter space  $\Phi$  is partitioned into two subspaces  $\Phi_1$  and  $\Phi_2$  such that decision  $\Phi_1$  is preferred if  $\theta \in \Phi_1$  and  $\Phi_2$  preferred if  $\theta \in \Phi_2$ . The costs of decisions  $\Phi_1$  and  $\Phi_2$  are, respectively,

1.1 
$$w_1(0)$$
  $\begin{cases} > 0, 0 \in \Omega_1 \\ > 0, 0 \in \Omega_2 \end{cases}$ 

and

1.2 
$$\mathbf{w}_{2}(\theta)$$
  $\begin{cases} \geqslant 0, \theta \in \Lambda_{1} \\ = 0, \theta \in \Lambda_{2} \end{cases}$ .

The motivating idea for Theorem A of this report is to be found in a lot acceptance sampling procedure proposed in an unpublished paper by Mr. Norman Rudy of Sacramento State College. Discussion with members of the Department of Statistics, Stanford University, in particular with Professor M. A. Girshick, was most beneficial in the formulation of the present draft of the report.

The main results of this paper will be based on the assumption that the probability distribution of x is continuous. The necessary modification of the results for the case of discrete probability distributions will be discussed in a concluding section of the paper.

If the parameter space consists of a simple dichotomy (i.e.  $\Omega_1$  consists of the single point  $\Theta_1$ , and  $\Omega_2$ , the single point  $\Theta_2$ , the minimax procedure is well known. The relevant statistic is the likelihood ratio

1.3 
$$z = z(\theta_1, \theta_2) = \frac{\prod_{j=1}^{n} p(x_1, \theta_2)}{\prod_{j=1}^{n} p(x_1, \theta_1)}$$
into the second second

and the decision procedure which minimizes the maximum expected cost is to

1.4 choose  $A_1$ , if  $\lambda \leqslant c$ , and

1.5 choose  $A_2$ , if  $\lambda > c$ ,

where the criterion c satisfies the relation

1.6 
$$w_1(\Theta_2) \Pr(\lambda \leq c|\Theta_2) = w_2(\Theta_1) \Pr(\lambda > c|\Theta_1)$$
.

This value of c is

1.7 
$$c = \frac{w_2(\theta_1)g}{w_1(\theta_2)(1-g)}$$
,

where g is the least favorable a priori probability that 0 = 01.

2. The Minimax Solution to a More General Problem. The assumption that  $\Omega_1$  and  $\Omega_2$  each consists of a single point is often a very artificial one. A more general formulation of the two-valued decision problem is the following:  $\Omega$  is an interval,  $\Theta \in \Omega_1$  if  $\Theta \leq \Theta_0$ , and  $\Theta \in \Omega_2$  if  $\Theta \gtrsim \Theta_0$ , where the costs of decisions  $A_1$  and  $A_2$ 

are given by 1.1, 1.2, and

2.1 
$$\mathbf{w}_{1}(\mathbf{e}_{0}) = \mathbf{w}_{2}(\mathbf{e}_{0}) = 0.2$$

This statement of the problem avoids the often unrealistic postulate of an "indifference interval", i.e. an interval such that decisions  $A_1$  and  $A_2$  are either costless or equally costly if 9 assumes a value in this interval. However, in this more general formulation, is the likelihood ratio test for some  $\Theta_1^{\varepsilon,\Omega}$  and  $\Theta_2^{\varepsilon,\Omega}$  still a minimax decision procedure?

The following theorem supplies the answer-

### Theorem A. Let

2.2 
$$R_1(c, \theta_1, \theta_2, \theta) = w_1(\theta) \Pr[\lambda(\theta_1, \theta_2) \le c/\theta]$$
 and

2.3 
$$R_2(c,\theta_1,\theta_2,\theta) = w_2(\theta) \Pr[\lambda(\theta_1,\theta_2) > c|\theta]$$
, where

 $\Theta_1 \in \Omega_1$  and  $\Theta_2 \in \Omega_2$ . Then the likelihood ratio test is a minimax procedure if and only if there exists a triple  $(c_0, \Theta_1^0, \Theta_2^0)$  such that

2.4 
$$\max_{Q \in \mathcal{N}_{2}} R_{1}(c_{0}, \Theta_{1}^{Q}, \Theta_{2}^{Q}, \Theta) = R_{1}(c_{0}, \Theta_{1}^{Q}, \Theta_{2}^{Q}, \Theta_{2}^{Q}) = R_{2}(c_{0}, \Theta_{1}^{Q}, \Theta_{2}^{Q}, \Theta_{1}^{Q})$$

$$= \max_{Q \in \mathcal{N}_{1}} R_{2}(c_{0}, \Theta_{1}^{Q}, \Theta_{2}^{Q}, \Theta) .$$

Proof of necessity:

Suppose the likelihood ratio test for some  $\theta_1^0 \in \Omega_1$ ,  $\theta_2^0 \in \Omega_2$ , and  $c = c^0$  is minimax. Then 2.4 follows.

It seems only reasonable to assume that  $\inf_{Q \in \Omega_2} w_1(Q) = w_1(Q)$  and  $\inf_{Q \in \Omega_2} w_2(Q) = w_2(Q)$ . With this in view, there is no loss in generality in requiring 2.1.

Proof of sufficiency:

Let  $F_0$  denote the likelihood ratio test procedure associated with the triple  $(c_0, \theta_1^0, \theta_2^0)$  satisfying 2.4. Then the expected cost of this procedure is given by

2.5 
$$R(F_0,G) = \int_{2}^{\pi_1(\theta)} \Pr[\lambda(\theta_1^0,\theta_2^0) \leq c_0[0]] dG(\theta)$$
  
+  $\int_{2}^{\pi_2(\theta)} \Pr[\lambda(\theta_1^0,\theta_2^0) > c_0[\theta]] dG(\theta)$ 

where dG(9) denotes any a priori probability measure over  $\Omega$ . Since  $R_1(c_0,\theta_1^0,\theta_2^0,\theta_2^0)$  and  $R_2(c_0,\theta_1^0,\theta_2^0,\theta_1^0)$  are the maxima of the integrands appearing in 2.5, it follows that

2.6 
$$R(F_0,G) \leqslant \int_{2}^{R_1(c_0,\theta_1^0,\theta_2^0,\theta_2^0)} dG(\theta) + \int_{1}^{R_2(c_0,\theta_1^0,\theta_2^0,\theta_1^0)} dG(\theta)$$

$$= R_1(c_0,\theta_1^0,\theta_2^0,\theta_2^0) = R_2(c_0,\theta_1^0,\theta_2^0,\theta_1^0) .$$

If G is a distribution concentrating all probability at  $\theta_1^0$  and  $\theta_2^0$ , 2.6 is a strict equality. Therefore

2.7 
$$\max_{G} R(F_0,G) = R_1(c_0,\theta_1^0,\theta_2^0,\theta_2^0) = R_2(c_0,\theta_1^0,\theta_2^0,\theta_1^0)$$

Let Go be the a priori distribution given by

$$g = Pr(\theta = 9_1^0)$$

where g satisfies

2.10 
$$c_0 = \frac{w_2(\theta_1^0)g}{w_1(\theta_2^0)(1-g)}$$
.

Clearly Fo is the Bayes procedure against Co. Therefore

2.11 
$$\min_{F} R(F_{\bullet}G_{0}) = R(F_{0},G_{0})$$
,

where F is any decision procedure for choosing between  $A_1$  and  $A_2$ . Since  $G_0$  is a distribution for which 2.6 is a strict equality, the minimax property of  $F_0$  follows.

3. Assumptions Under Which the Likelihood Ratio Test Is a Minimax Procedure. It is now of interest to examine a class of distribution functions and a class of cost functions for which condition 2.4 of Theorem A is satisfied.

Consider a family of probability measures defined by

3.1 
$$dL_0(\pi/\theta) = \frac{1}{\omega_0(\theta)} e^{\Theta x} d\psi_0(x)$$
,

where  $\psi_0(x)$  is a measure defined over an interval X, and where

3.2 
$$\omega_o(\theta) = \int_{-\infty}^{\infty} e^{9x} dY_o(x)$$

is a bounded function of  $\theta$  defined over an interval  $\Omega_{\theta}$  If u is the sum of n independent random variables each distributed according to 3.1, then u is distributed according to

The properties of this family of distribution functions are given an extensive discussion in Section 3 of Girshick and Savage, "Bayes and Minimax Estimates for Quadratic Loss Functions", Technical Report No. 5, Contract No. N6-ONR-251 Task III, Department of Statistics, Stanford University, November 21, 1950.

3.3 
$$dL(u|\theta) = \frac{1}{\omega(\theta)} e^{\theta u} dY(u)$$
.

This is again the distribution family introduced in 3.1 and 3.2, with Y(u) defined over an interval U.

In view of the form of the distributions 3.1 and 3.3, it is simple to show that for every number c such that  $z \le c$  (where z is defined in 1.3) there exists a unique number k such that  $z \le c$  This correspondence permits the functions 2.2 and 2.3 to be rewritten as

3.4 
$$R_1(k,\theta) = w_1(\theta) L(k|\theta)$$

3.5 
$$R_2(k,\theta) = w_2(\theta) [1-L(k|\theta)]$$
.

These functions have the following important property:

#### Lemma 1.

If  $\max_{\theta \in \Lambda_2} R_1(k,\theta)$  and  $\max_{\theta \in \Lambda_2} R_2(k,\theta)$  exist, then they are, respectively,  $\max_{\theta \in \Lambda_2} R_1(k,\theta)$  monotonic increasing and decreasing functions of k.

Proof:

Let 
$$\max_{\theta \in \Omega_2} R_1(k,\theta) = w_1(\theta^*)L(k|\theta^*), \theta^* \in \Omega_2$$

and 
$$\max_{Q \in \Lambda_1} R_2(k, \Theta) = w_2(\Theta^n)[1-L(k|\Theta^n)], \Theta^n \in \Lambda_1$$
.

Obviously  $L(k|\theta)$  is an increasing function of k for any  $\theta$ . Thus for  $\Delta$  positive

$$w_{1}(\theta) L(k|\theta) \leq w_{1}(\theta) L(k+\Delta|\theta) \leq \max_{\theta \in \Omega_{S}} R_{1}(k+\Delta,\theta)$$
 and

$$w_2(\theta^n)[1-L(k|\theta^n)] \leq w_2(\theta^n)[1-L(k-\Delta|\theta^n)] \leq \max_{\theta \in \Omega_1} w_2(\theta)[1-L(k-\Delta|\theta^n)]$$

$$\equiv \max_{0 \in A_1} R_2(k-A,0),$$

as desired.

It is now possible to prove the following

Theorem B. Let  $\psi(u)$ ,  $w_1(\theta)$ , and  $w_2(\theta)$  be continuous throughout

any finite intervals in U = (a,b),  $\Omega_2$ , and  $\Omega_1$ , respectively. 4

3.6 
$$\lim_{k \to a} R_1(k,\theta) = \lim_{k \to b} R_2(k,\theta) = 0$$

uniformly in 0. Then if the  $\max_{\theta \in \Omega} R_1(k,\theta)$  and  $\max_{\theta \in \Omega} R_2(k,\theta)$  exist for some keV, there exists a unique value of keV such that

3.7 
$$\max_{\Theta \in \Omega_2} R_1(k,\Theta) = \max_{\Theta \in \Omega_1} R_2(k,\Theta).$$

Proof:

Suppose for some  $k = k_0 \in U$ ,  $\max_{Q \in \Omega_2} R_1(k_0, 0) > \max_{Q \in \Omega_2} R_2(k_0, 0)$ . For  $Q \in \Omega_2$  is uniformly bounded by  $\max_{Q \in \Omega_2} R_1(k_0, 0)$ . Since  $R_1(k_0, 0)$  is continuous in k and 0, the  $\max_{Q \in \Omega_2} R_1(k_0, 0)$  exists and is continuous for  $k < k_0$ . Certainly the  $\sup_{Q \in \Omega_2} R_2(k_0, 0)$  is continuous in k, and for  $k < k_0$  sup  $R_2(k_0, 0) \ge \max_{Q \in \Omega_1} R_2(k_0, 0)$ . Since by 3.6  $\max_{Q \in \Omega_2} R_2(k_0, 0) \to 0$  as  $\max_{Q \in \Omega_1} R_2(k_0, 0) = \max_{Q \in \Omega_2} R_2(k$ 

The interval U may of course be the open interval  $(-\infty, \infty)$ .

Corollary 1. Let  $\Psi(u)$ ,  $w_1(\theta)$ , and  $w_2(\theta)$  be continuous and bounded over U = (a,b),  $\Omega_2$ , and  $\Omega_1$ , respectively. 5/ Then there exists a unique keU such that  $\max_{\theta \in \Omega_2} R_2(k,\theta)$ .

Corollary 2. Let 
$$\delta_{2^{Q}}$$
 3.8  $w_{1}(9) = 0(e^{-\delta_{1}Q})$   $\delta_{2} > 0$  3.9  $w_{2}(9) = 0(e^{-\delta_{1}Q})$   $\delta_{1} > 0$ 

be continuous functions throughout any finite intervals in  $\Omega_2$  and  $\Omega_1$ , respectively. Let  $\Psi(u)$  be continuous, positive, and less than  $\Psi(\infty)$  for any real u. Then there exists a unique  $k \in U$  such that  $\max_{\theta \in \Omega_2} R_1(k,\theta) = \max_{\theta \in \Omega_1} R_2(k,\theta)$ .

Proof:

For  $\Delta > 0$ ,

(1) 
$$\lim_{\theta\to\infty} w_1(\theta)L(k|\theta) \le \lim_{\theta\to\infty} e^{\delta_2\theta} \frac{\int_0^k e^{\theta u} d\psi(u)}{\int_0^\infty e^{\theta u} d\psi(u)}$$
. constant

$$\leq \lim_{\theta \to \infty} \frac{\int_{0}^{k} e^{(u+\delta_2)\theta} d\Psi(u)}{\int_{0}^{\infty} e^{\theta u} d\Psi(u)} \cdot constant$$

$$k+\delta_2+\Delta$$

$$\leq \lim_{\theta \to \infty} \frac{e^{(\delta_2 + k)\theta}}{e^{(\delta_2 + \Delta + k)\theta}} = \frac{\int_0^k d\gamma(u)}{\int_0^\infty d\gamma(u)} \cdot \text{constant } = 0, \text{ and }$$

$$k + \delta_2 + \Delta$$

 $<sup>\</sup>frac{5}{\gamma(u)}$  is of course always bounded.

(2) 
$$\lim_{\theta \to -\infty} w_2(\theta) [1-L(k|\theta)] \leq \lim_{\theta \to -\infty} e^{\int_0^\infty e^{\theta u} d\gamma(u)}$$
 constant

$$\leq \lim_{\theta \to -\infty} \frac{k}{k^{-\delta_1 - \Delta}} e^{u\theta} dV(u)$$
 constant

$$\leq \lim_{\Theta \to -\infty} \frac{e^{(k-\delta_1)\Theta} \int_{0}^{\infty} d\gamma(u)}{e^{(k-\delta_1-\Delta)\Theta} \int_{-\infty}^{k-\delta_1-\Delta} d\gamma(u)} \cdot constant$$

= 0 .

Since

(3) 
$$\lim_{\theta \to \infty} w_1(\theta) L(k|\theta) = \lim_{\theta \to -\infty} w_2(\theta) [1-L(k|\theta)] = w_1(\theta_0) L(k|\theta_0)$$
  
=  $w_2(\theta_0) [1-L(k|\theta_0)] = 0$ 

for any k, the max  $R_1(k,\theta)$  and max  $R_2(k,\theta)$  exist for any k.  $\theta \epsilon n_1$ 

The min  $\omega(\theta)$  and min  $\omega(\theta)$  are positive constants. Therefore,  $\theta \epsilon \Omega_2$  for  $\theta \epsilon \Omega_2$ ,

$$(4) \quad \mathbf{w_1}(\Theta) \mathbf{L}(\mathbf{k}|\Theta) \leq e^{\delta_2 \Theta} \int_{-\infty}^{\mathbf{k}} e^{\Theta \mathbf{u}} \, dY(\mathbf{u}). \quad \text{constant}$$

$$\leq \int_{-\infty}^{\mathbf{k}} e^{(\delta_2 + \mathbf{u})(\Theta - \Theta_0 + 1)} e^{(\delta_2 + \mathbf{u})(\Theta_0 - 1)} \, dY(\mathbf{u}). \quad \text{constant}$$

$$\leq e^{(\delta_2 + \mathbf{k})(\Theta - \Theta_0 + 1)} \int_{-\infty}^{\mathbf{k}} e^{(\delta_2 + \mathbf{u})(\Theta_0 - 1)} \, dY(\mathbf{u}) \cdot \text{constant}$$

$$\leq e^{\delta_2 + \mathbf{k}} \int_{-\infty}^{\mathbf{k}} e^{(\delta_2 + \mathbf{u})(\Theta_0 - 1)} \, dY(\mathbf{u}) \cdot \text{constant},$$

and similarly, for  $\Theta \in \Omega_{\gamma - S}$ 

(5) 
$$w_2(0)[1-L(k|0)] \le e^{k-\delta_1} \int_{k}^{\infty} e^{(u-\delta_1)(\Omega_0+1)} dV(u)$$
 constant.

The inequalities (4) and (5) demonstrate that 3.6 of Theorem B is satisfied. The conclusion of Corollary 2 follows.

Theorem B serves primarily as a criterion for determining whether or not minimax livelihood ratio tests exist in particular cases. The corollaries to Theorem B are interesting as applications of this criterion in the analysis of certain classes of sampling distributions and cost functions. Individual cases could of course be subjected to a much more direct analysis.

4. Remarks on the Discrete Case. The generality of the family of distributions introduced in 3.1 should not be overlooked: it includes many of the most important distributions encountered in statistics, such as the normal, binomial, and Poisson distributions. Suppose the distribution under consideration in this class

is a discrete one, and suppose that Y(u) assumes jumps at each value of a denumerable sequence in which the values are placed in order of increasing magnitude. For example, in the binomial distribution,  $u = 0,1,2,\ldots,n$ . In general it will not be possible to find a value of k in such a sequence so that  $\max_{\theta \in \Omega_2} R_2(k,\theta) = \max_{\theta \in \Omega_2} R_2(k,\theta)$ .

However, provided that all other conditions of Theorem B are satisfied except for the continuity of  $\Psi(u)$ , it will be possible to find a pair  $(k_0, k_0)$  such that

4.1 
$$\max_{\Theta \in \Omega_2} R_1(k_0 \Theta) < \max_{\Theta \in \Omega_1} R_2(k_0, \Theta)$$

4.2 
$$\max_{\theta \in \Lambda_2} R_1(k_0', \theta) > \max_{\theta \in \Lambda_1} R_2(k_0', \theta) ,$$

where  $k_0^{\dagger}$  is the next larger value in the sequence than  $k_0$ . The minimax procedure is then a convex linear mixture of two procedures involving  $k_0$  and  $k_0^{\dagger}$ , a mixture determined by

4.3 
$$\max_{\theta \in \Omega_2} [fR_1(k_0, \theta) + (1-f)R_1(k_0, \theta)] = \max_{\theta \in \Omega_1} [fR_2(k_0, \theta) + (1-f)R_2(k_0, \theta)]_0$$

where O<f<1.

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